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```
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%
% Project Code: YPEA120
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%
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%
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%

clc;
clear;
close all;
```

Problem Definition

```
CostFunction = @(x) MOP4(x);          % Cost Function

nVar = 3;                             % Number of Decision Variables

VarSize = [1 nVar];                   % Size of Decision Variables Matrix

VarMin = -5;                          % Lower Bound of Variables
VarMax = 5;                           % Upper Bound of Variables

% Number of Objective Functions
nObj = numel(CostFunction(unifrnd(VarMin, VarMax, VarSize)));
```

Testing

```
% Independent variable
popVals = [40 80 120 160 200];        % Population Size
kungs = 1; % whether to use naive (0) or kung's (1) method
```

```

numTrials = 200;      % trials per scenario

% Statistics
executiontime = zeros(2, width(popVals), numTrials);
avgTime = zeros(2, width(popVals));

% algorithm parameters
MaxIt = 10;          % Maximum Number of Iterations
pCrossover = 0.7;    % Crossover Percentage
pMutation = 0.4;     % Mutation Percentage
mu = 0.02;           % Mutation Rate
sigma = 0.1*(VarMax-VarMin); % Mutation Step Size
for kungs = 0:1
    for pIdx = 1:width(popVals)
        nPop = popVals(pIdx);      % Population Size
        nCrossover = 2*round(pCrossover*nPop/2); % Number of Parents
        (Offsprings)
        nMutation = round(pMutation*nPop); % Number of Mutants
        for t = 1:numTrials
            % Initialization
            empty_individual.Position = [];
            empty_individual.Cost = [];
            empty_individual.Rank = [];
            empty_individual.DominationSet = [];
            empty_individual.DominatedCount = [];
            empty_individual.CrowdingDistance = [];
            pop = repmat(empty_individual, nPop, 1);
            for i = 1:nPop
                pop(i).Position = unifrnd(VarMin, VarMax, VarSize);
                pop(i).Cost = CostFunction(pop(i).Position);
            end

            % Non-Dominated Sorting
            [pop, F] = NonDominatedSorting(pop);

            % Calculate Crowding Distance
            pop = CalcCrowdingDistance(pop, F);

            % Sort Population
            [pop, F] = SortPopulation(pop);

            % NSGA-II Main Loop
            tic;
            for it = 1:MaxIt
                % Crossover
                popc = repmat(empty_individual, nCrossover/2, 2);
                for k = 1:nCrossover/2
                    i1 = randi([1 nPop]);
                    p1 = pop(i1);
                    i2 = randi([1 nPop]);
                    p2 = pop(i2);
                    [popc(k, 1).Position, popc(k, 2).Position] =
Crossover(p1.Position, p2.Position);

```

```

1).Position);
popc(k, 2).Cost = CostFunction(popc(k,
2).Position);
end
popc = popc(:);
% Mutation
popm = repmat(empty_individual, nMutation, 1);
for k = 1:nMutation
    i = randi([1 nPop]);
    p = pop(i);
    popm(k).Position = Mutate(p.Position, mu, sigma);
    popm(k).Cost = CostFunction(popm(k).Position);
end

% Merge
pop = [pop
       popc
       popm]; %#ok

% algorithm
if kungs
    % sort by 1st objective
    carr = [pop.Cost];
    [~, idx] = sort(carr(1,:), 2, 'ascend');
    pop = pop(idx);

    % determine fronts
    idx = 1:height(pop);
    fr = 1;
    while size(idx) > 0
        [~, F{fr}] = Front(pop(idx), idx);
        for fi = F{fr}
            pop(fi).Rank = fr;
        end
        idx = setdiff(idx, F{fr});
        fr = fr + 1;
    end

    % truncate
    pop = CalcCrowdingDistance(pop, F);
    [pop, F] = SortPopulation(pop);
    pop = pop(1:nPop);

    % sort once more
    [pop, F] = SortPopulation(pop);

    % store F1
    F1 = pop(F{1});
else
    % Non-Dominated Sorting
    [pop, F] = NonDominatedSorting(pop);

    % Calculate Crowding Distance

```

```

        pop = CalcCrowdingDistance(pop, F);

        % Sort Population
        pop = SortPopulation(pop);

        % Truncate
        pop = pop(1:nPop);

        % Non-Dominated Sorting
        [pop, F] = NonDominatedSorting(pop);

        % Calculate Crowding Distance
        pop = CalcCrowdingDistance(pop, F);

        % Sort Population
        [pop, F] = SortPopulation(pop);

        % Store F1
        F1 = pop(F{1});
    end

    % Show Iteration Information
    %disp(['Iteration ' num2str(it) ': Number of F1
Members = ' num2str(numel(F1))]);

    % Plot F1 Costs
    %figure(1);
    %PlotCosts(F1);
    %pause(0.01);
end
    executiontime(kungs+1, pIdx, t) = toc;
end
    disp(['Completed nPop=' num2str(nPop) ' with kungs='
num2str(kungs)]);
end
end

Completed nPop=40 with kungs=0
Completed nPop=80 with kungs=0
Completed nPop=120 with kungs=0
Completed nPop=160 with kungs=0
Completed nPop=200 with kungs=0
Completed nPop=40 with kungs=1
Completed nPop=80 with kungs=1
Completed nPop=120 with kungs=1
Completed nPop=160 with kungs=1
Completed nPop=200 with kungs=1

```

Results

```

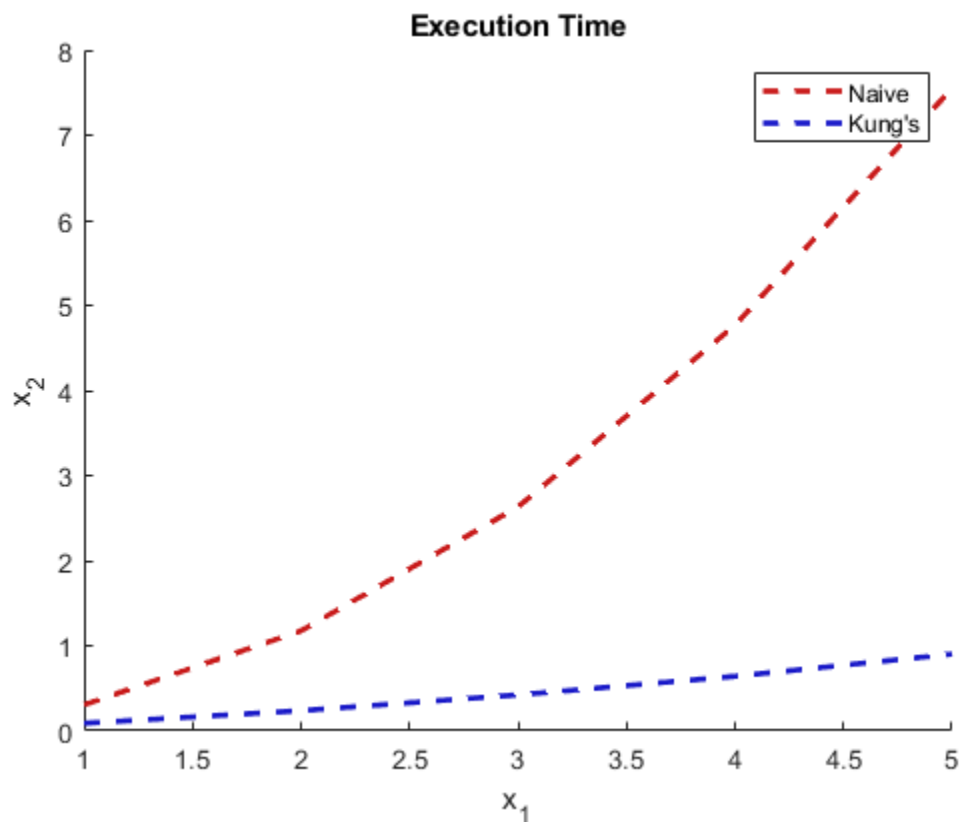
for p = 1:width(popVals)
    avgTime(1,p) = mean(executiontime(1,p,:));
    avgTime(2,p) = mean(executiontime(2,p,:));

```

```

end
figure();
hold on;
title('Execution Time');
xlabel('x_1');
ylabel('x_2');
ax = gca;
ax.XAxisLocation='origin';
ax.YAxisLocation='origin';
x1 = 1:width(popVals);
plot(x1, avgTime(1,:), 'Color',
[.8 .1 .1], 'LineStyle', '--', 'LineWidth', 2, 'DisplayName', 'Naive');
plot(x1, avgTime(2,:), 'Color',
[.1 .1 .8], 'LineStyle', '--', 'LineWidth', 2, 'DisplayName', "Kung's");
legend();

```



Comparison

Kung's method scales significantly better than the naive implementation. At a lower population size the difference isn't as significant, but the scaling of the naive implementation fails relatively quickly. Execution time thus skyrockets, while Kung's method stays relatively the same. This aligns with the statement that Naive is $O(n^2)$ and Kung's is $O(n \log(n))$. Kung's, however, might use more memory. Since matlab is a pass-by-value language, every recursion of Kung's gets its own copy of the data. This probably ends up being more than the single copy of *pop* that the naive implementation creates. However, I have no way of accurately measuring this, so I can only theorize. Regardless, the extra memory is absolutely worth the increase in execution speeds, which becomes roughly tenfold for the population size of 200.

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